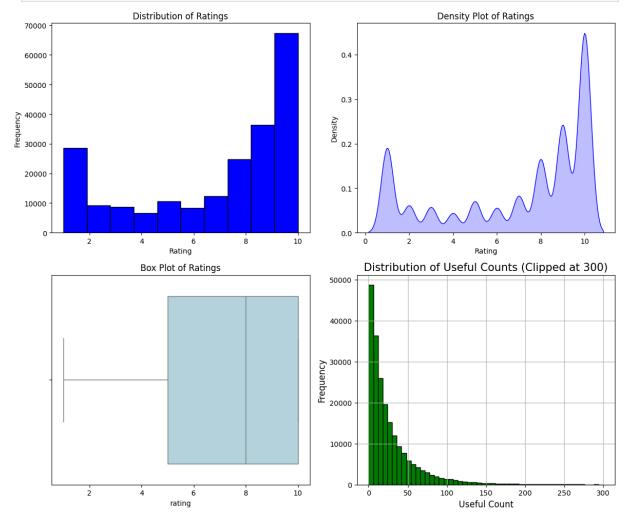
```
In [1]: # Imports
        import matplotlib.pyplot as plt
        import matplotlib.dates as mdates
        import seaborn as sns
        import numpy as np
        import pandas as pd
        from wordcloud import WordCloud
        from ucimlrepo import fetch ucirepo
        pd.options.mode.chained assignment = None
In [2]: # fetch dataset from the UCI archive website by its ID.
        drug dataset id = 462
        Drugs = fetch ucirepo(id=drug dataset id)
In [3]: # variable info.
        print(Drugs.metadata["additional info"]["variable info"])
       1. drugName (categorical): name of drug
       2. condition (categorical): name of condition
       3. review (text): patient review
       4. rating (numerical): 10 star patient rating
       5. date (date): date of review entry
       6. usefulCount (numerical): number of users who found review useful
In [4]: # Create a total data from the Drugs
        data original = Drugs.data.original
        # Checking for missing values in the dataset
        missing values = data original.isnull().sum()
        # Drop the rows with the missing condition value.
        data original = data original.dropna()
        # Drop the rows with an improper condition. rows with </span> do not include
        span count = data original['condition'].str.contains('</span>', na=False).st
        data_original = data_original[~data_original['condition'].str.contains('</sr</pre>
        # Basic summary statistics for numerical columns
        numerical summary = data original.describe()
        # Distribution of unique values for categorical columns
        drug count = data original['drugName'].nunique()
        condition count = data original['condition'].nunique()
        # Summarize text review column (average length of reviews)
        data original['review length'] = data original['review'].apply(lambda x: ler
        average review length = data original['review length'].mean()
        # Display key stats
        print(f"Missing Values \n{missing values}\n")
        print(f'Number of rows with "</span>": {span count}')
        print(f"Numerical Summary: \n{numerical summary}\n")
```

```
print(f"Number of unique drugs: {drug count}")
        print(f"Number of unique conditions: {condition count}")
        print(f"Average length of review: {round(average review length)}")
      Missing Values
       id
                         0
      drugName
                        0
       condition
                      1194
       review
                        0
       rating
                        0
      date
                         0
       usefulCount
                         0
      dtype: int64
      Number of rows with "</span>": 1171
      Numerical Summary:
                                              usefulCount
                         id
                                    rating
       count 212698.000000 212698.000000 212698.000000
      mean
             116079.388673
                                  6.992431
                                               28.186819
       std 66999.171961
                                  3.275994
                                               36.455651
      min
                   0.000000
                                 1.000000
                                               0.000000
      25%
             58154.250000
                                  5.000000
                                                6.000000
      50% 115961.500000
                                 8.000000
                                               16.000000
      75%
             174009.750000
                                10.000000
                                               37.000000
             232291.000000
                                10.000000
                                             1291.000000
      max
      Number of unique drugs: 3654
      Number of unique conditions: 836
      Average length of review: 85
In [5]: # Convert 'date' column to datetime format
        data original['date'] = pd.to datetime(data original['date'], format="%d-%b-
        # Extract year from the date
        data original['year'] = data original['date'].dt.year
        # Extract the month from the date
        data original['month'] = data original['date'].dt.month
        # Group data by year and month, and calculate the number of reviews and aver
        data original['year month'] = data original['date'].dt.to period('M')
In [6]: # Histogram for numerical variables (rating and usefulCount, assuming they \epsilon
        fig, axs = plt.subplots(2, 2, figsize=(12, 10))
        axs[0, 0].hist(data_original["rating"], bins=10, edgecolor="black", color="t
        axs[0, 0].set title('Distribution of Ratings')
        axs[0, 0].set xlabel('Rating')
        axs[0, 0].set ylabel('Frequency')
        sns.kdeplot(data original['rating'], fill=True, color='blue', ax=axs[0, 1])
        axs[0, 1].set title('Density Plot of Ratings')
        axs[0, 1].set xlabel('Rating')
        axs[0, 1].set ylabel('Density')
        plt.grid(True)
        # Box plots to show spread and outliers
```

```
sns.boxplot(x=data_original["rating"], color="lightblue", ax=axs[1, 0])
axs[1, 0].set_title("Box Plot of Ratings")

# Distribution of Useful counts plot
axs[1, 1].hist(data_original['usefulCount'], bins=50, color='green', edgecol
axs[1, 1].set_title('Distribution of Useful Counts (Clipped at 300)', fontsi
axs[1, 1].set_xlabel('Useful Count', fontsize=12)
axs[1, 1].set_ylabel('Frequency', fontsize=12)
axs[1, 1].grid(True)

plt.tight_layout()
plt.show()
```



In [7]: # Calculate the average rating for each condition
 avg\_rating\_per\_condition = data\_original.groupby("condition")["rating"].mear
 condition\_review\_counts = data\_original.groupby("condition").size() # Count

conditions\_at\_least\_1000\_reviews = condition\_review\_counts[condition\_review\_filtered\_avg\_ratings = avg\_rating\_per\_condition.loc[conditions\_at\_least\_1006]

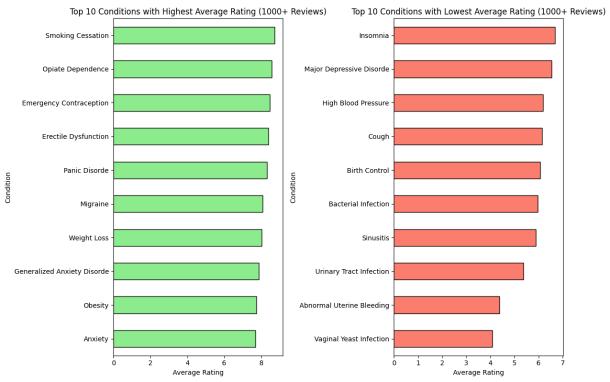
# Sort the conditions by average rating
sorted\_ratings = filtered\_avg\_ratings.sort\_values()

# Visualize the top 10 highest and lowest rated conditions with at least 106
plt.figure(figsize=(12, 8))

```
# Plot the top 10 highest rated conditions
plt.subplot(1, 2, 1)
sorted_ratings.tail(10).plot(kind="barh", color="lightgreen", edgecolor="bla
plt.title("Top 10 Conditions with Highest Average Rating (1000+ Reviews)")
plt.xlabel("Average Rating")
plt.ylabel("Condition")

# Plot the top 10 lowest rated conditions
plt.subplot(1, 2, 2)
sorted_ratings.head(10).plot(kind="barh", color="salmon", edgecolor="black")
plt.title("Top 10 Conditions with Lowest Average Rating (1000+ Reviews)")
plt.xlabel("Average Rating")
plt.ylabel("Condition")

# Display the plots
plt.tight_layout()
plt.show()
```



```
In [8]: # string arr of all of the months in a year, used for the x ticks
months = ['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug', 'Sep', 'Oc
#Including the effect of time aka including the date information
avg_rating_per_drug = data_original.groupby("drugName")["rating"].mean()
drug_review_counts = data_original.groupby("drugName").size()

drugs_at_least_1000_reviews = drug_review_counts[drug_review_counts >= 1000]
filtered_avg_drug_ratings = avg_rating_per_drug.loc[drugs_at_least_1000_revi

# Sort the conditions by average rating
sorted_drug_ratings = filtered_avg_drug_ratings.sort_values()

# Identify the top 10 most used drugs (based on number of reviews)
top_10_drugs = sorted_drug_ratings.tail(10).index
```

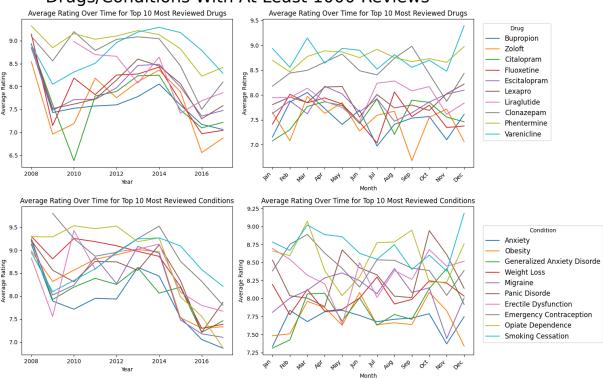
```
top 10 conditions = sorted ratings.tail(10).index
# Group by drug and year/month, then calculate the average rating
ratings by year top drugs = data original[data original['drugName'].isin(top
ratings by month top drugs = data original[data original['drugName'].isin(to
# Group by condition and year/month, then calculate the average rating
ratings by year top conditions = data original[data original['condition'].is
ratings by month top conditions = data original[data original['condition'].i
# Create a 2x2 grid of subplots
fig, axs = plt.subplots(2, 2, figsize=(12, 10))
fig.suptitle('Drugs/Conditions With At Least 1000 Reviews', fontsize=30)
# Plotting for each drug and condition
for i in range(len(top 10 drugs)):
    drug = top 10 drugs[i]
    sns.lineplot(x='year', y='rating', data=ratings_by_year_top_drugs[rating
    sns.lineplot(x='month', y='rating', data=ratings by month top drugs[rati
    condition = top 10 conditions[i]
    sns.lineplot(x='year', y='rating', data=ratings by year top conditions[r
    sns.lineplot(x='month', y='rating', data=ratings by month top conditions
axs[0, 0].set title('Average Rating Over Time for Top 10 Most Reviewed Drugs
axs[0, 0].set xlabel('Year')
axs[0, 0].set ylabel('Average Rating')
axs[0, 1].set title('Average Rating Over Time for Top 10 Most Reviewed Drugs
axs[0, 1].set xlabel('Month')
axs[0, 1].set ylabel('Average Rating')
axs[0, 1].set xticks(range(1, 13)) # Set ticks for each month (1 to 12)
axs[0, 1].set xticklabels(months)
axs[1, 0].set title('Average Rating Over Time for Top 10 Most Reviewed Condi
axs[1, 0].set xlabel('Year')
axs[1, 0].set ylabel('Average Rating')
axs[1, 1].set title('Average Rating Over Time for Top 10 Most Reviewed Condi
axs[1, 1].set xlabel('Month')
axs[1, 1].set ylabel('Average Rating')
axs[1, 1].set xticks(range(1, 13)) # Set ticks for each month (1 to 12)
axs[1, 1].set xticklabels(months)
# Collect handles and labels from one of the subplots
handles drugs, labels drugs = axs[0, 0].get legend handles labels()
handles condition, labels condition = axs[1, 0].get legend handles labels()
# Remove the legends as it will be displayed on the side
axs[0, 0].legend().remove()
axs[0, 1].legend().remove()
axs[1, 0].legend().remove()
axs[1, 1].legend().remove()
```

```
# Set a common legend outside the plot (to the right)
fig.legend(handles_drugs, labels_drugs, loc='center left', bbox_to_anchor=(1
fig.legend(handles_condition, labels_condition, loc='center left', bbox_to_a

# Rotate the x-ticks to make them readable
plt.setp(axs[0, 1].get_xticklabels(), rotation=45, ha="right")
plt.setp(axs[1, 1].get_xticklabels(), rotation=45, ha="right")

plt.tight_layout()
plt.show()
```

# Drugs/Conditions With At Least 1000 Reviews

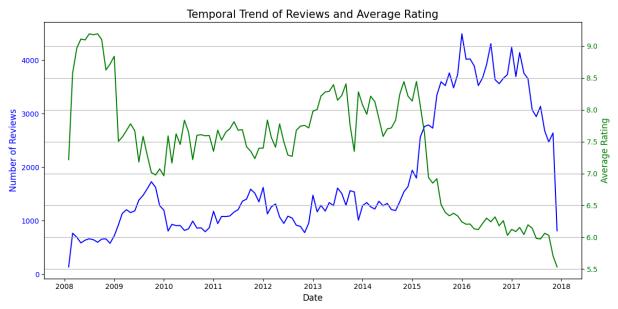


```
In [9]: # Aggregating data by month (number of reviews and average rating)
        temporal data = data original.groupby('year month').agg(
            num_reviews=('review', 'count'),
            avg rating=('rating', 'mean')
        ).reset index()
        # Convert year month back to datetime for proper plotting
        temporal data['year month'] = temporal data['year month'].dt.to timestamp()
        # Plotting temporal trends
        fig, ax1 = plt.subplots(figsize=(12, 6))
        # Plot number of reviews over time
        ax1.plot(temporal data['year month'], temporal data['num reviews'], color='t
        ax1.set_xlabel('Date', fontsize=12)
        ax1.set ylabel('Number of Reviews', fontsize=12, color='blue')
        ax1.tick params(axis='y', labelcolor='blue')
        # Create a second y-axis for the average rating
        ax2 = ax1.twinx()
        ax2.plot(temporal data['year month'], temporal data['avg rating'], color='gr
```

```
ax2.set_ylabel('Average Rating', fontsize=12, color='green')
ax2.tick_params(axis='y', labelcolor='green')

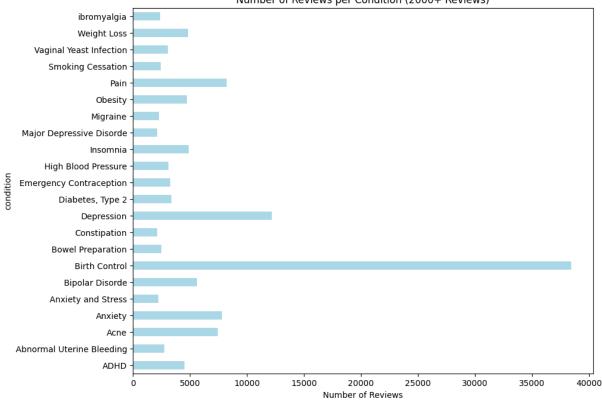
plt.title('Temporal Trend of Reviews and Average Rating', fontsize=15)
plt.grid(True)
plt.tight_layout()

plt.show()
```

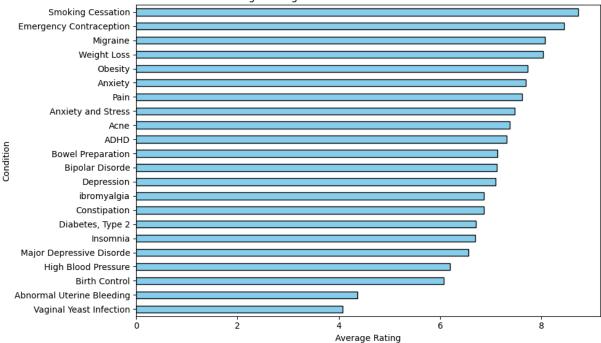


```
In [10]: conditions_large_reviews = condition_review_counts[condition_review_counts >
    plt.figure(figsize=(12, 5))
    conditions_large_reviews.plot(kind="barh", figsize=(10, 8), color="lightblue plt.title("Number of Reviews per Condition (2000+ Reviews)")
    plt.xlabel("Number of Reviews")
    plt.show()
```

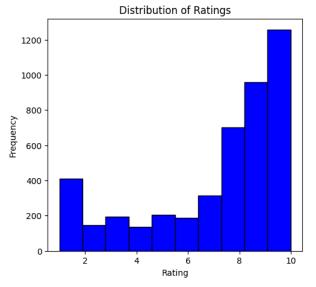








```
In [12]: adhd = data_original[data_original["condition"] == "ADHD"]
         plt.figure(figsize=(12, 5))
         # Histogram for "rating"
         plt.subplot(1, 2, 1)
         plt.hist(adhd["rating"], bins=10, color="blue", edgecolor="black")
         plt.title("Distribution of Ratings")
         plt.xlabel("Rating")
         plt.ylabel("Frequency")
         # Generate a string of all drug names repeated based on their counts
         drug_counts = adhd["drugName"].head(30).value_counts()
         drug string = " ".join([f"{drug} " * count for drug, count in drug_counts.it
         # Create the word cloud
         wordcloud = WordCloud(width=1000, height=800, background color="white").gene
         # Plot the word cloud
         plt.subplot(1, 2, 2)
         plt.imshow(wordcloud, interpolation="bilinear")
         plt.axis("off") # Turn off the axis
         plt.title("Word Cloud of Drugs by Number of Reviews")
         plt.show()
         # Display the plots
         plt.tight layout()
         plt.show()
```



Daytrana
Intuniv Methylphenidate
dextroamphetamine
Lisdexamfetamine

Amphetamine

Amphetamine

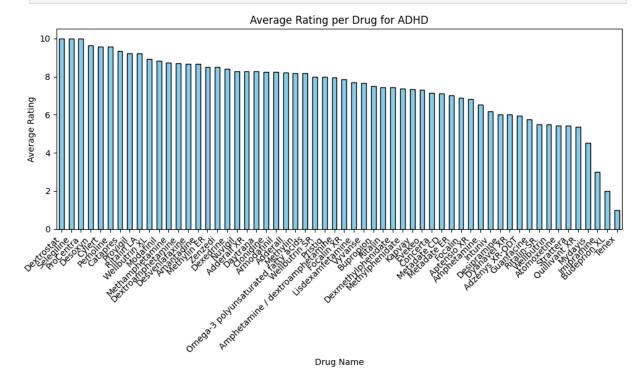
Concerta Vyvanse
GuanfacineBupropion
XR
Focalin

<Figure size 640x480 with 0 Axes>

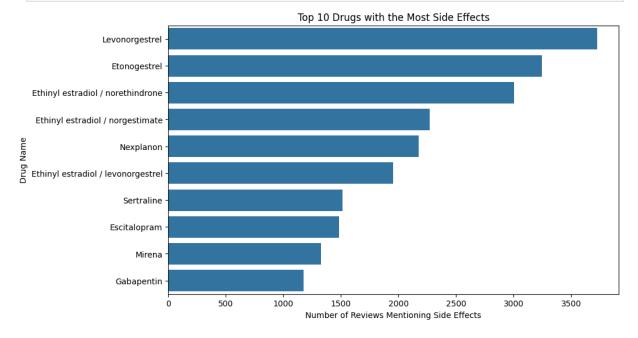
```
In [13]: # Calculate the average rating for each drug
    avg_rating_per_drug = adhd.groupby("drugName")["rating"].mean()

# Plotting the average rating per drug
    plt.figure(figsize=(10, 6))
    avg_rating_per_drug.sort_values(ascending=False).plot(kind="bar", color="sky
    plt.title("Average Rating per Drug for ADHD")
    plt.xlabel("Drug Name")
    plt.ylabel("Average Rating")
    plt.xticks(rotation=45, ha="right")
    plt.tight_layout()

# Show the plot
    plt.show()
```



```
# Import local module helper function that checks if side effects are mention
from utils.side effects import check side effects
# Apply the function to the review text and create the 'side effects' column
data original['side effects'] = data original['review'].apply(check side eff
# Check the distribution of the new target variable
side effect counts = data original['side effects'].value counts()
#print(f"Side Effect Target Distribution:\n{side effect counts}")
# Group by drug name and sum the 'side effects' column to count how many rev
side effects by drug = data original.groupby('drugName')['side effects'].sum
# Sort the results to find the druas with the most side effects
side effects by drug sorted = side effects by drug.sort values(by='side effe
# Plot the top 10 drugs with the most side effects
plt.figure(figsize=(10, 6))
sns.barplot(x='side_effects', y='drugName', data=side_effects_by_drug_sorted
plt.title('Top 10 Drugs with the Most Side Effects')
plt.xlabel('Number of Reviews Mentioning Side Effects')
plt.ylabel('Drug Name')
plt.show()
```



```
In [15]: from nltk.sentiment.vader import SentimentIntensityAnalyzer
from sklearn.model_selection import train_test_split
sid = SentimentIntensityAnalyzer()

# Split the data into 75% training and 25% testing
train_df, test_df = train_test_split(data_original, test_size=0.25, random_s
# Preprocess text for sentiment analysis (get sentiment score from review te
data_original['sentiment'] = data_original['review'].apply(lambda x: sid.pol

# Separate the data into two groups: negative (sentiment < 0) and neutral/po
negative = data_original[data_original['sentiment'] < 0]
neutral_positive = data_original[data_original['sentiment'] >= 0]
```

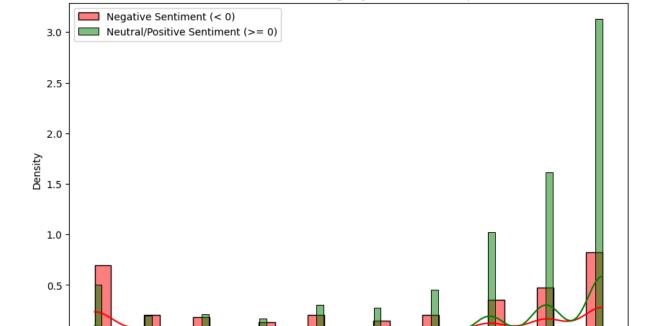
```
avg_rating_negative = negative['rating'].mean()
avg_rating_neutral_positive = neutral_positive['rating'].mean()

# Print the results
print(f"Average rating for negative sentiment (sentiment < 0): {avg_rating_r
print(f"Average rating for neutral/positive sentiment (sentiment >= 0): {avg

# plot the distribution of ratings for each sentiment group
plt.figure(figsize=(10, 6))
sns.histplot(negative['rating'], color='red', label='Negative Sentiment (< 6
sns.histplot(neutral_positive['rating'], color='green', label='Neutral/Posit
plt.title('Distribution of Ratings by Sentiment Groups')
plt.xlabel('Rating')
plt.ylabel('Density')
plt.legend()
plt.show()</pre>
```

Distribution of Ratings by Sentiment Groups

Average rating for negative sentiment (sentiment < 0): 6.08 Average rating for neutral/positive sentiment (sentiment >= 0): 7.94



Rating

```
In [16]: from sklearn.preprocessing import LabelEncoder
    from sklearn.ensemble import RandomForestClassifier
    from sklearn.metrics import accuracy_score, classification_report

# Define effectiveness for training data, here we assumed the ratings > 5 me
# which can be adjusted withe the previous section of sentiment + rating
    train_df['effective'] = train_df.apply(
        lambda row: 1 if row['rating'] > 5 else 0, axis=1
)

# Define effectiveness for test data in the same way
    test_df['effective'] = test_df.apply(
```

```
lambda row: 1 if row['rating'] > 5 else 0, axis=1
# Handle missing data
train df.dropna(subset=['drugName', 'condition', 'rating'], inplace=True)
test df.dropna(subset=['drugName', 'condition', 'rating'], inplace=True)
# Encode categorical variables (Drug name, Condition)
le drug = LabelEncoder()
le condition = LabelEncoder()
train df['drug encoded'] = le drug.fit transform(train df['drugName'])
train df['condition encoded'] = le condition.fit transform(train df['conditi
# Define a safe transformation function to handle unseen labels
def safe transform(encoder, data, default value=-1):
    return np.array([default value if label not in encoder.classes else end
# Apply safe transformation on the test data to handle unseen drugs and cond
test df['drug encoded'] = safe transform(le drug, test df['drugName'])
test df['condition encoded'] = safe transform(le condition, test df['conditi
# Define the input features (drug name and condition) and the target variabl
X train = train df[['drug encoded', 'condition encoded']]
y train = train df['effective']
X test = test df[['drug encoded', 'condition encoded']]
y test = test df['effective']
# Train a Random Forest classifier using drug name and condition as input f\epsilon
clf = RandomForestClassifier(n estimators=100, random state=42)
clf.fit(X train, y train)
# Make predictions on the test set using drug name and condition
y pred = clf.predict(X test)
# Evaluate the model
print(f"Train Accuracy: {accuracy score(y train, clf.predict(X train))}")
print(f"Test Accuracy: {accuracy score(y test, y pred)}")
print(classification report(y test, y pred))
# Create a comparison table of actual vs predicted values
comparison df = pd.DataFrame({
    'Drug Name': test df['drugName'],
    'Condition': test df['condition'],
    'Actual Effectiveness': y test,
    'Predicted Effectiveness': y pred
})
print(comparison df.head())
# Plot the predicted effectiveness for a few drug-condition pairs
plt.figure(figsize=(10, 6))
sns.countplot(x='Actual Effectiveness', hue='Predicted Effectiveness', data=
plt.title('Comparison of Actual vs Predicted Effectiveness')
plt.show()
```

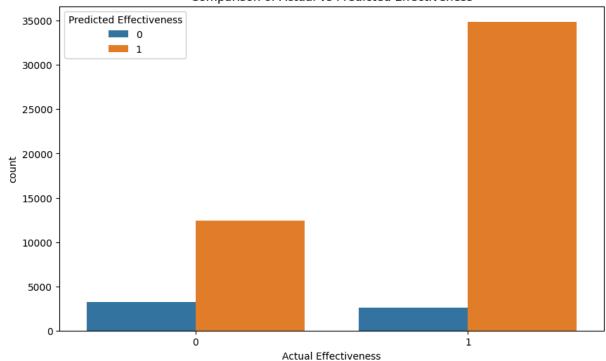
Train Accuracy: 0.7290484757683845 Test Accuracy: 0.7173859896567936

	precision	recall	f1-score	support
0	0.56 0.74	0.21 0.93	0.30 0.82	15697 37478
accuracy macro avg weighted avg	0.65 0.68	0.57 0.72	0.72 0.56 0.67	53175 53175 53175

	Drug Name	Condition \
212040	Ethinyl estradiol / etonogestrel	Birth Control
177292	Dextromethorphan	Cough
117623	Topamax	Migraine Prevention
32740	Phentermine / topiramate	Weight Loss
193096	Vasotec	Left Ventricular Dysfunction

	Actual Effectiveness	Predicted	Effectiveness
212040	0		1
177292	0		0
117623	0		1
32740	1		1
193096	1		1

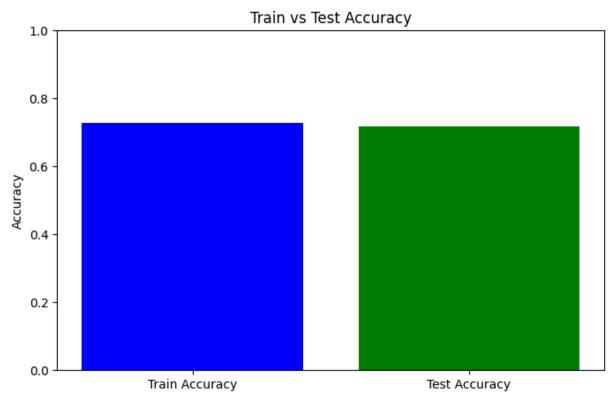
Comparison of Actual vs Predicted Effectiveness

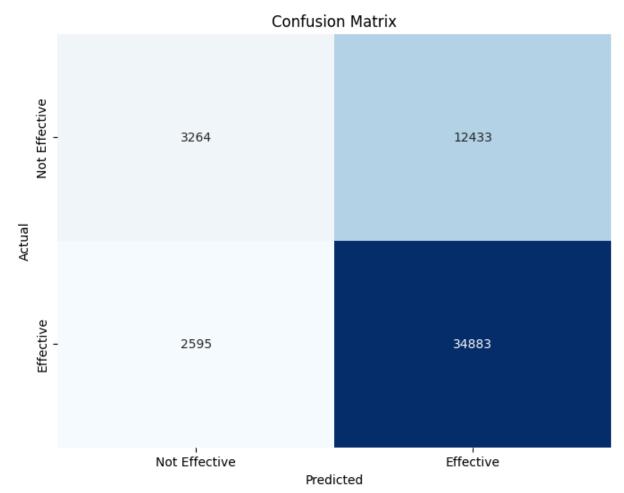


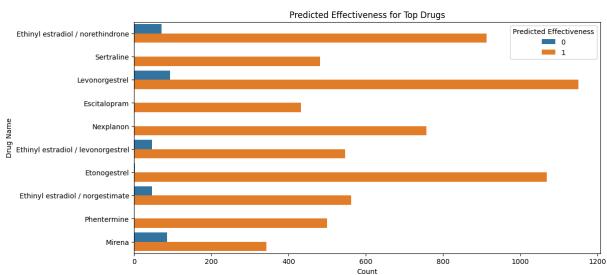
```
In [17]: #Some visualizations, we can pick and chose what we want to keep
from sklearn.metrics import confusion_matrix
import seaborn as sns

# Train and Test Accuracy as a bar plot
plt.figure(figsize=(8, 5))
accuracy_values = [accuracy_score(y_train, clf.predict(X_train)), accuracy_s
plt.bar(['Train Accuracy', 'Test Accuracy'], accuracy_values, color=['blue',
```

```
plt.title('Train vs Test Accuracy')
plt.ylabel('Accuracy')
plt.ylim(0, 1)
plt.show()
# Confusion Matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False, xticklabels=[
plt.title('Confusion Matrix')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
# Bar plot of Predicted Effectiveness for different drugs
plt.figure(figsize=(12, 6))
top_drugs = comparison_df['Drug Name'].value_counts().index[:10]
sns.countplot(y='Drug Name', hue='Predicted Effectiveness', data=comparison
plt.title('Predicted Effectiveness for Top Drugs')
plt.xlabel('Count')
plt.ylabel('Drug Name')
plt.legend(title='Predicted Effectiveness', loc='upper right')
plt.show()
```







```
In [18]: # Some additional descriptive analysis that could be useful:
    # Select all numerical columns except 'sentiment'
    numerical_columns = data_original.select_dtypes(include=['float64', 'int64']
    numerical_columns = numerical_columns.drop('sentiment', errors='ignore') #

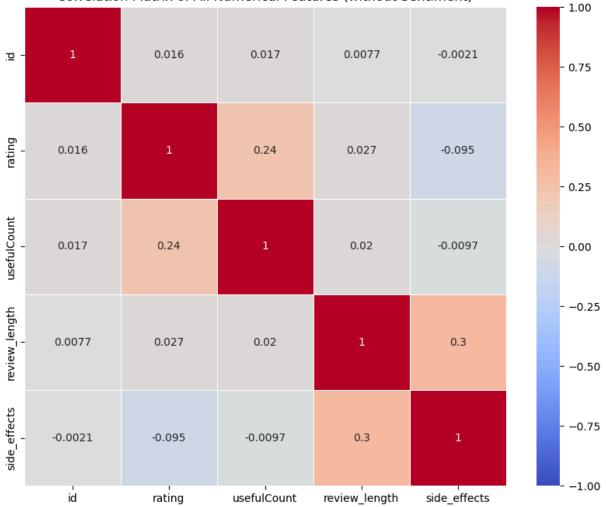
# Calculate the correlation matrix for the selected numerical features
    correlation_matrix = data_original[numerical_columns].corr()

# Display the correlation matrix
    print("Correlation Matrix (without Sentiment):")
```

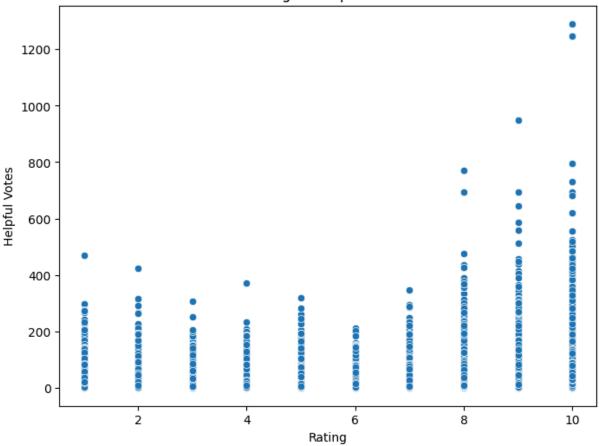
```
print(correlation matrix)
 # Plot a heatmap of the correlation matrix
 plt.figure(figsize=(10, 8))
 sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', vmin=-1, vmax=1
 plt.title('Correlation Matrix of All Numerical Features (without Sentiment)'
 plt.show()
 # Scatter plot for rating vs usefulCount (helpful votes)
 plt.figure(figsize=(8, 6))
 sns.scatterplot(x='rating', y='usefulCount', data=data_original)
 plt.title('Rating vs Helpful Votes')
 plt.xlabel('Rating')
 plt.ylabel('Helpful Votes')
 plt.show()
Correlation Matrix (without Sentiment):
```

	id	rating	usefulCount	review_length	side_effects
id	1.000000	0.016090	0.017253	0.007651	-0.002111
rating	0.016090	1.000000	0.235375	0.027199	-0.094916
usefulCount	0.017253	0.235375	1.000000	0.019819	-0.009713
review_length	0.007651	0.027199	0.019819	1.000000	0.303048
side_effects	-0.002111	-0.094916	-0.009713	0.303048	1.000000





## Rating vs Helpful Votes

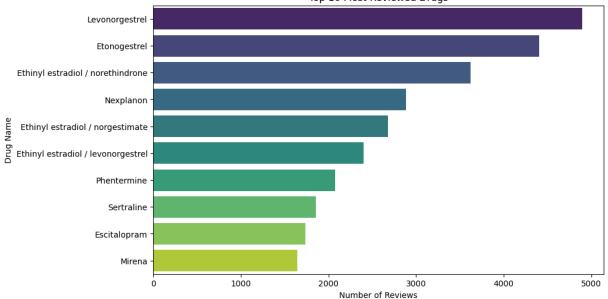


```
In [19]: # Top 10 most common drugs
         top drugs = data original['drugName'].value counts().head(10)
         print("Top 10 Most Reviewed Drugs:")
         print(top drugs)
         # Visualize the top 10 most common drugs
         plt.figure(figsize=(10, 6))
         sns.barplot(x=top drugs.values, y=top drugs.index, hue=top drugs.index, lege
         plt.title('Top 10 Most Reviewed Drugs')
         plt.xlabel('Number of Reviews')
         plt.ylabel('Drug Name')
         plt.show()
         # Top 10 most common conditions
         top conditions = data original['condition'].value counts().head(10)
         print("Top 10 Most Reviewed Conditions:")
         print(top conditions)
         # Visualize the top 10 conditions
         plt.figure(figsize=(10, 6))
         sns.barplot(x=top conditions.values, y=top conditions.index, hue=top conditi
         plt.title('Top 10 Most Reviewed Conditions')
         plt.xlabel('Number of Reviews')
         plt.ylabel('Condition')
         plt.show()
         # Number of reviews per drug and condition
```

```
reviews per drug condition = data original.groupby(['drugName', 'condition']
 top drug condition = reviews per drug condition.sort values(by='num reviews'
 print("Top 10 Drug-Condition Pairs by Number of Reviews:")
 print(top drug condition)
 # Visualize the top drug-condition pairs
 plt.figure(figsize=(10, 6))
 sns.barplot(x='num reviews', y='drugName', hue='condition', data=top drug cd
 plt.title('Top 10 Drug-Condition Pairs by Number of Reviews')
 plt.xlabel('Number of Reviews')
 plt.ylabel('Drug Name')
 plt.legend(title='Condition', bbox to anchor=(1.05, 1), loc='upper left')
 plt.show()
 # Plot the distribution of sentiment scores
 plt.figure(figsize=(10, 6))
 sns.histplot(data original['sentiment'], bins=20, kde=True, color='purple')
 plt.title('Sentiment Score Distribution')
 plt.xlabel('Sentiment Score')
 plt.ylabel('Count')
 plt.show()
 # Show how many reviews have positive, negative, or neutral sentiment
 print("Sentiment Distribution:")
 sentiment labels = ['Negative', 'Neutral', 'Positive']
 sentiment distribution = [
     (data original['sentiment'] < 0).sum(),</pre>
     (data original['sentiment'] == 0).sum(),
     (data original['sentiment'] > 0).sum()
 for label, count in zip(sentiment labels, sentiment distribution):
     print(f"{label}: {count} reviews")
Top 10 Most Reviewed Drugs:
drugName
                                      4896
```

Levonorgestrel Etonogestrel 4402 Ethinyl estradiol / norethindrone 3619 Nexplanon 2883 Ethinyl estradiol / norgestimate 2682 Ethinyl estradiol / levonorgestrel 2400 Phentermine 2077 Sertraline 1859 Escitalopram 1739 1647 Mirena Name: count, dtype: int64



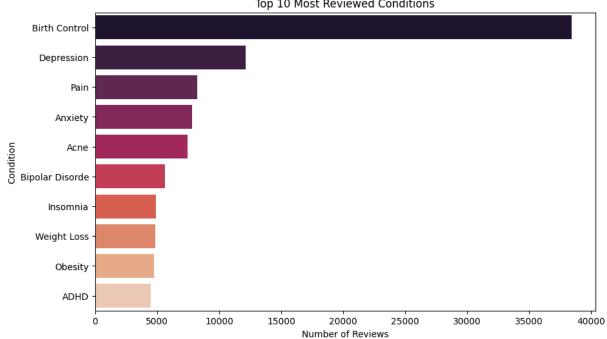


Top 10 Most Reviewed Conditions:

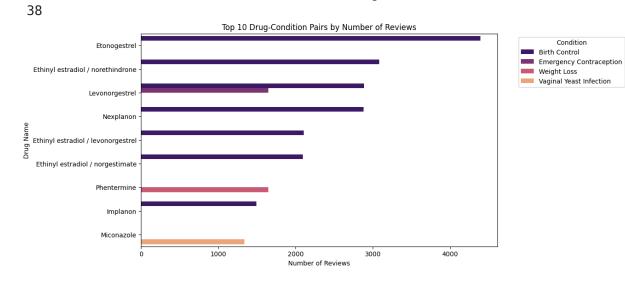
condition

Birth Control 38436 Depression 12164 Pain 8245 7812 Anxiety Acne 7435 Bipolar Disorde 5604 Insomnia 4904 Weight Loss 4857 Obesity 4757 ADHD 4509 Name: count, dtype: int64

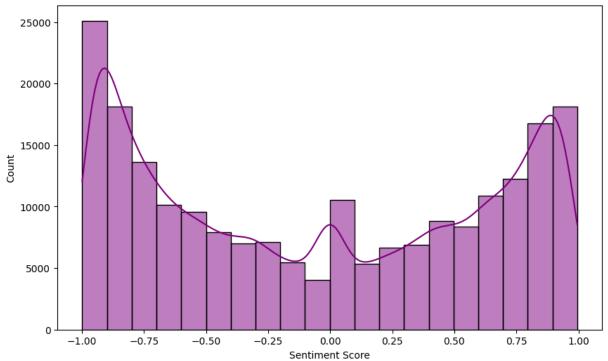
Top 10 Most Reviewed Conditions



Top 10	O Drug-Condition Pairs by Number of drugName	Reviews: condition	num_revie
WS			
3073	Etonogestrel	Birth Control	43
94 3049 81	Ethinyl estradiol / norethindrone	Birth Control	30
4442	Levonorgestrel	Birth Control	28
84 5520	Nexplanon	Birth Control	28
83 3040	Ethinyl octradial / layanargactral	Birth Control	21
07	Ethinyl estradiol / levonorgestrel	BIT CHI CONCLOC	21
3058	Ethinyl estradiol / norgestimate	Birth Control	20
97	1	Emanage Control of the	1.0
4443 51	Levonorgestrei	Emergency Contraception	16
6136	Phentermine	Weight Loss	16
50			
3943	Implanon	Birth Control	14
96	Micanorala	Vacinal Vacat Infaction	10
5099	MICONAZOLE	Vaginal Yeast Infection	13



### Sentiment Score Distribution



Sentiment Distribution: Negative: 108118 reviews Neutral: 6691 reviews Positive: 97889 reviews

```
In [20]: import ipywidgets as widgets
         from IPython.display import display
         # Define effectiveness using percentiles of the ratings to add more variance
         # Compute percentiles and scale effectiveness
         data original['percentile effectiveness'] = data original.groupby('condition')
         # Group by condition and drug, then calculate the mean percentile effectiver
         effectiveness by condition = data original.groupby(['condition', 'drugName']
             average rating=('rating', 'mean'),
             confidence=('percentile effectiveness', 'mean') # Now using percentile
         ).reset index()
         # Function to get the best drug for a given condition
         def get best drug(condition):
             condition data = effectiveness by condition[effectiveness by condition['
             if condition data.empty:
                 return "No data available", 0
             best drug = condition data.sort values(by='confidence', ascending=False)
             return best drug['drugName'], best drug['confidence'] # Return the conf
         # Create a dropdown widget for selecting a condition
         condition dropdown = widgets.Dropdown(
             options=data original['condition'].unique(),
             description='Condition:',
```

```
disabled=False
)

# Create a label widget to display the best drug and confidence
result_label = widgets.Label(value="Select a condition to see the best drug"

# Function to update the result when a condition is selected
def on_condition_change(change):
    if change['type'] == 'change' and change['name'] == 'value':
        selected_condition = change['new']
        best_drug, confidence = get_best_drug(selected_condition)
        result_label.value = f"Best_Drug: {best_drug}\nConfidence: {confider

# Attach the function to the dropdown widget
condition_dropdown.observe(on_condition_change)

# Display the dropdown and result label in the notebook
display(condition_dropdown)
display(result_label)
```

Dropdown(description='Condition:', options=('Left Ventricular Dysfunction',
'ADHD', 'Birth Control', 'Opiate D...
Label(value='Select a condition to see the best drug')

This notebook was converted with convert.ploomber.io